

Improving the PIC method: Moment-preserving particle resampling and higher order velocity-space particle-grid mapping in the global total-f gyrokinetic particle-in-cell code XGC

A. Mollén¹, V. Carey², M. F. Adams³, M. G. Knepley⁴,
R. Hager¹, J. Dominski¹ and C. S. Chang¹ +
XGC/WDMAApp team

¹ Princeton Plasma Physics Laboratory, Princeton, NJ, USA

² Theodon Consulting LLC Superior, CO, USA

³ Lawrence Berkeley National Laboratory, Berkeley, CA, USA

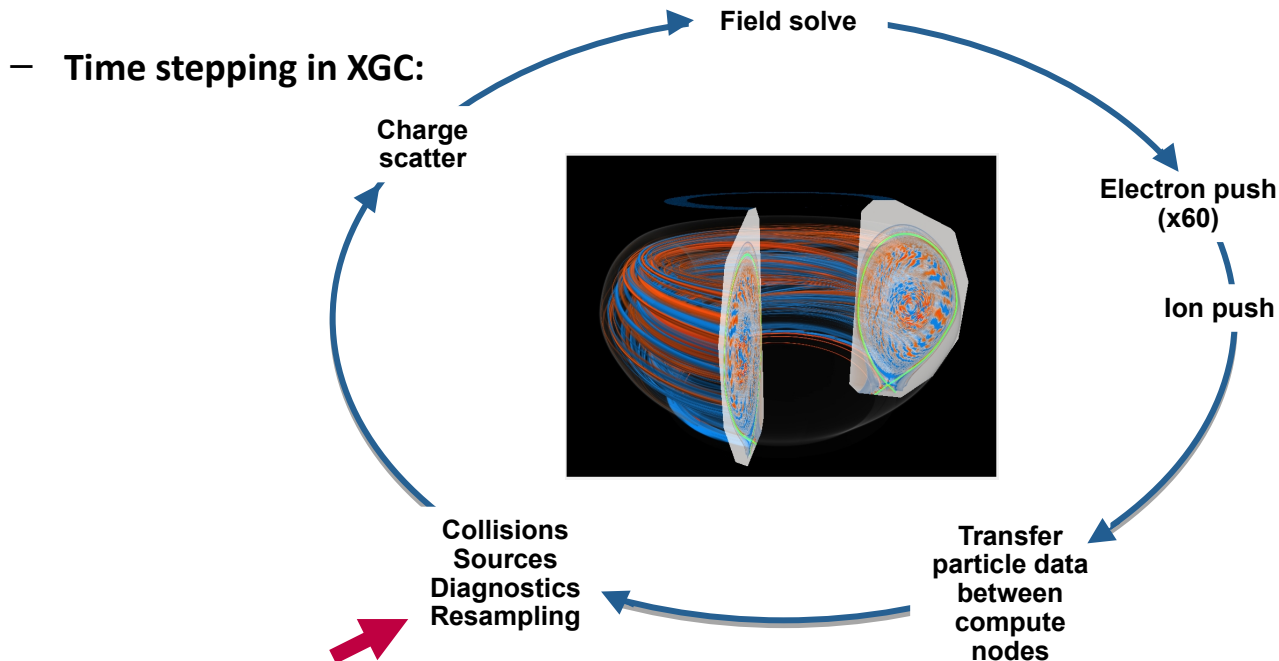
⁴ State University of New York at Buffalo, NY, USA

- Motivation
- Brief introduction to the XGC code
- Particle resampling in a PIC code
- Updated velocity-space interpolation using a pseudo-inverse
- Summary

- Main target of the XGC code is to study boundary plasmas that may contain a steep edge pedestal.
- XGC is being coupled to the core codes GENE and GEM in the ECP-WDMAApp project.
- Here we discuss two significant mathematical progresses that allow pedestal simulation and code coupling to be more accurate and efficient:
 - Moment-preserving particle resampling technique to reduce the load-imbalance and enhance simulation accuracy.
 - A novel velocity-space interpolation using the pseudo-inverse and particle resampling methods to remove the numerical heating in the steep edge pedestal.

- Motivation
- Brief introduction to the XGC code
- Particle resampling in a PIC code
- Updated velocity-space interpolation using a pseudo-inverse
- Summary

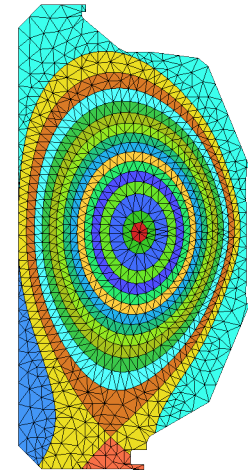
- The global total-f gyrokinetic particle-in-cell (PIC) code XGC xgc.pppl.gov
 - Study transport in magnetic fusion plasmas.
 - Neoclassical and turbulent dynamics + neutral particle transport.
 - Treats full plasma volume including scrape-off layer across separatrix.
 - Code has been converted from Fortran to C++ with Kokkos/Cabana for GPUs.



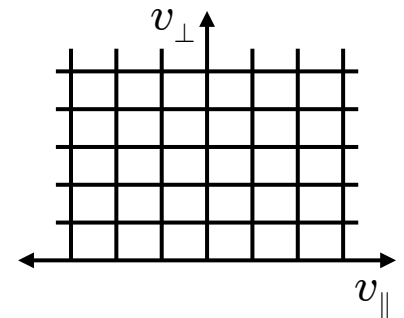
Focus of this presentation

- Typical size for XGC turbulence simulations
 - 0.1M- 1M configuration-space vertices per poloidal plane.
 - $N_{\text{plane}} \sim 16 - 128$.
 - 10 000 particles/vertex: trillions of total particles for ITER.
 - Velocity space grid: $N_{v_{||}} \times N_{v_{\perp}} \sim 40 \times 40$.
- Why Coulomb collisions on 5D grid?
 - Cost of particle-particle collision scheme $\geq O(N_{\text{particles}} \text{ or } N_{\text{particles}}^2)$.
 - Cost of grid-based scheme $\sim O(N_{\text{vertices}}) + \text{interpolation cost}$.
 $N_{\text{vertices}} \ll N_{\text{particles}} \rightarrow$ much cheaper.
 - The grid is also used for coupling to other codes in the Exascale Computing Project.
- Exact energy conservation (avoiding numerical heating) is difficult to obtain in particle-velocity grid coupling (interpolation). Other schemes can face similar issues [Alves et al, Phys. Rev. E (2021)].

Unstructured triangular mesh
in 3D configuration space



Uniform rectangular grid
in 2D velocity space



- The total distribution function of species s is divided as [\[Ku et al, PoP \(2018\)\]](#)

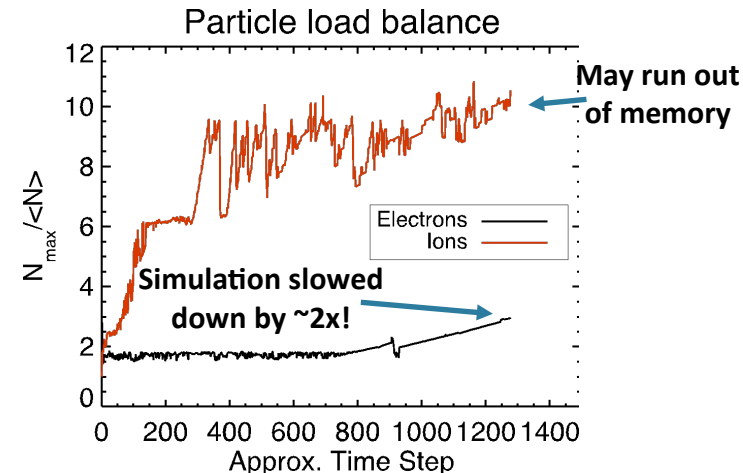
$$f_s = f_0 + f_p = f_a + f_g + f_p$$

- Slowly varying part $f_0 = f_a + f_g$
 - f_a analytic function (Maxwellian)
 - f_g represented on the 5D phase-space continuum grid
- Fast-varying physics (particle orbital, instability and turbulent dynamics)
 - f_p represented by marker particles, with two weights:
time-independent full-f weight and time dependent weight
- No restriction on relative magnitude of f_p compared with f_0 .
- A fraction of f_p can be transferred over to f_g and f_a during a time step to control particle weight growth.

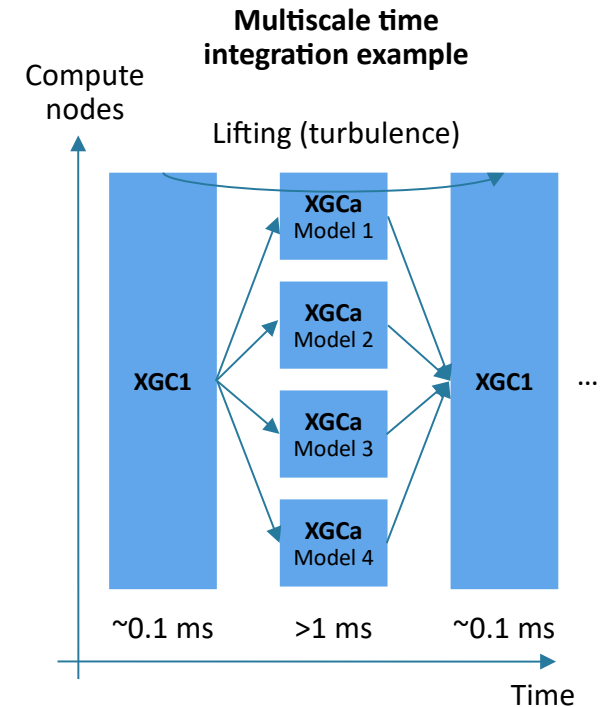
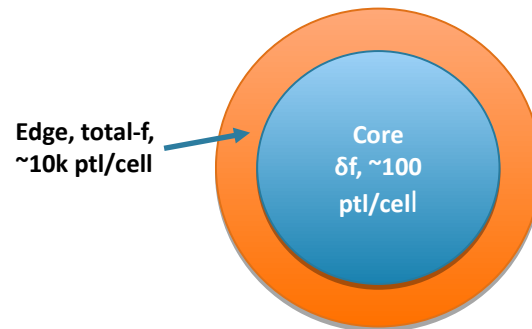
- Motivation
- Brief introduction to the XGC code
- Particle resampling in a PIC code
- Updated velocity-space interpolation using a pseudo-inverse
- Summary

- Particle-in-cell codes have advantages, e.g.
 - Easily parallelizable and extremely scalable.
 - Good for high-dimensional problems and complicated boundary conditions.
- There are disadvantages for a long-time simulation
 - Time dependent sampling noise \rightarrow error accumulation.
 - Complicated load-balancing in parallel simulation.
- Particle resampling can improve both accuracy and load balancing
 - Ensure good particle coverage in phase-space.
 - Homogenize the particle workload over compute nodes.

\rightarrow Allows longer time simulation.



- Load balancing
- Noise reduction (noise = variance in marker particle weights)
- Enable long time-scale simulation
- Quiet start (mitigate initial transient from particle loading)
- Multiscale time integration
combining full-scale turbulence simulation with
axisymmetric simulation and anomalous transport model
- Core-edge coupling of δf (core) and total-f (edge) code
[\[Dominski et al, PoP \(2021\)\]](#)

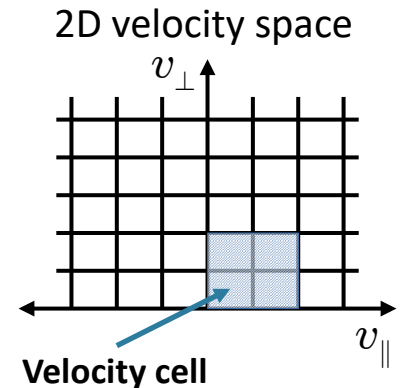
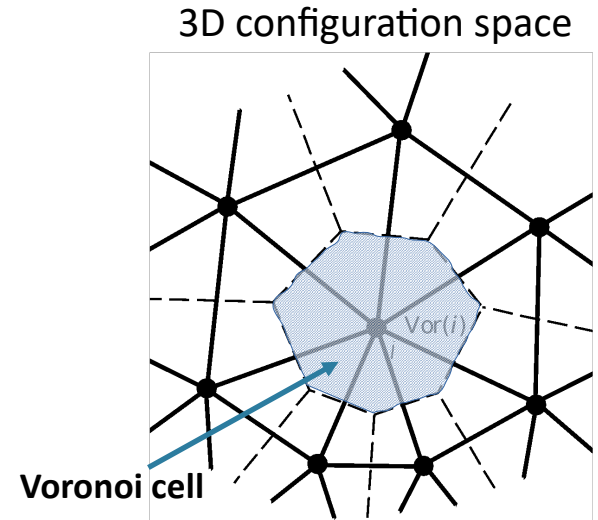


- **Velocity space mapping with pseudo-inverse**

Moment-preserving particle resampling [Faghihi et al, JCP (2020)]

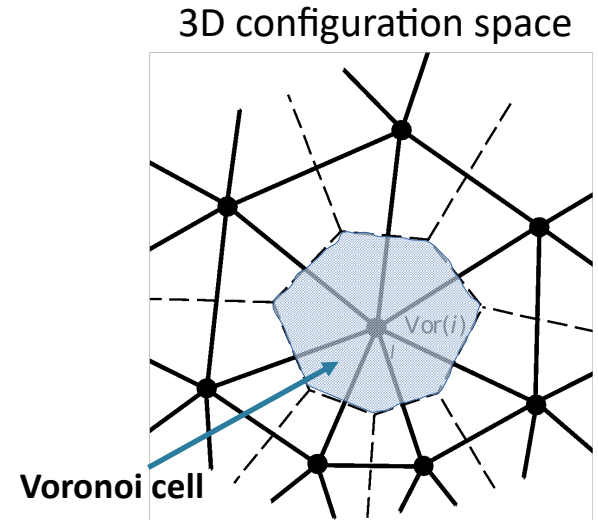
- Sort marker particles into phase-space bins
 - Voronoi cells from 3D mesh nodes + halfway between poloidal planes.
 - 2D velocity space cells.
- Up/down/re-sample particles in each bin
 - Add, remove or maintain particles in the bin.
 - Make adjustments to ensure that the total phase-space volume of the particles in every bin is preserved.
- Compute new particle weights through variance minimization subject to a set of constraints e.g. mass, momentum, kinetic energy conservation.
- Optimization is solved using constrained quadratic programming

$$\begin{aligned} & \text{minimize} \quad \frac{1}{2} \mathbf{w}^T \mathbf{Q} \mathbf{w} + \mathbf{c}^T \mathbf{w} \\ & \text{subject to} \quad \begin{pmatrix} A_1 \\ A_2 \end{pmatrix} \mathbf{w} \begin{cases} = b_1 \\ \leq b_2 \end{cases} \end{aligned}$$

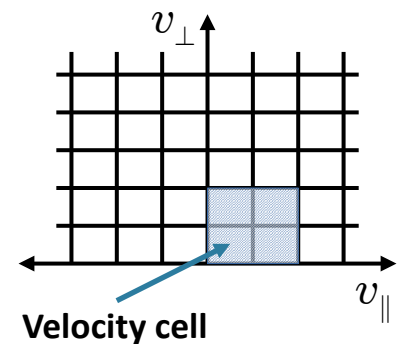


Moment-preserving particle resampling [Faghihi et al, JCP (2020)]

- There are several degrees of freedom (room for optimization)
 - Definition of phase-space bins.
 - Choice of moments to preserve, e.g. avoid ill-conditioned matrices from similar constraints.
 - How to sample marker particles, e.g. from which distribution.
 - How to add/remove marker particles in up/down-sampling.
- The resampling works well for the typical use case of every 10-20 time step
 - All bins are independent so perfectly parallelizable.
 - Good performance with combined process and thread parallelism (MPI + OpenMP).
- Future work
 - Convert module from Fortran to C++.
 - Put on GPUs.
 - Machine learning.



2D velocity space



- In some special cases we want to perform aggressive upsampling, but this seems to cause a numerical instability.

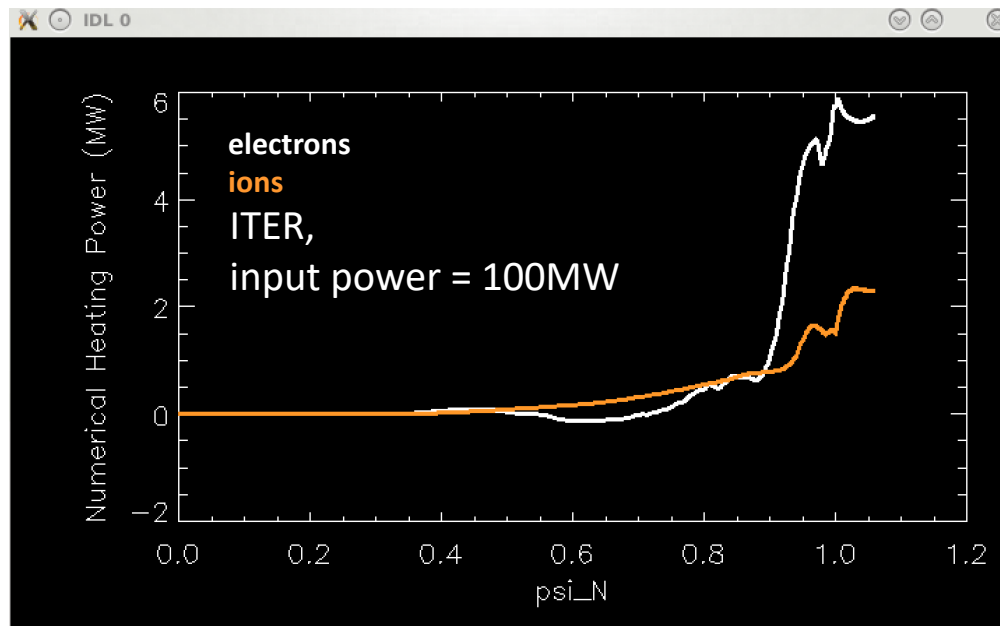
Show animation of time evolution of perturbed potential without vs with aggressive resampling

- Possible solutions
 - Marker particle placement in bins.
 - Choice of moments to preserve and how to sample marker particles.
 - Nonlinear programming (expensive).

- Motivation
- Brief introduction to the XGC code
- Particle resampling in a PIC code
- Updated velocity-space interpolation using a pseudo-inverse
- Summary

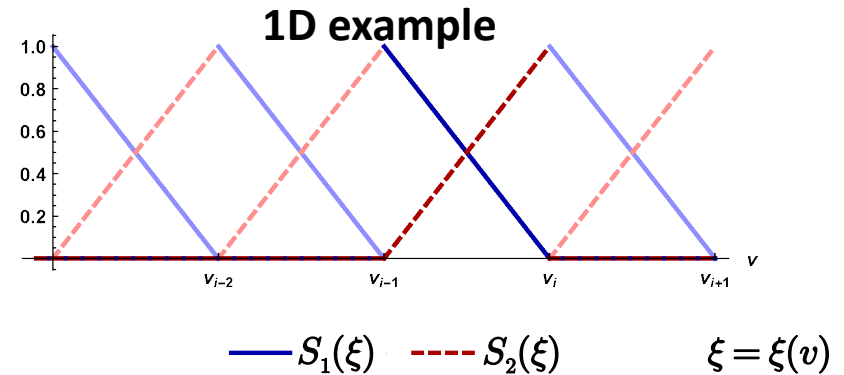
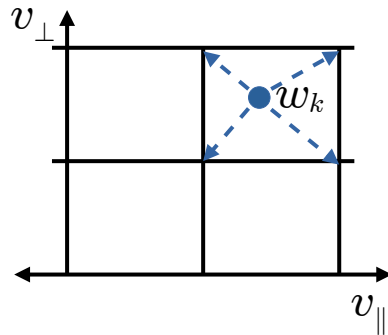
- Dissipative operations (collisions) and sources on 5D grid can have arbitrarily small conservation errors by stricter tolerances in iterative solvers.
- Bilinear interpolation error can be controlled by increasing the number of marker particles
[Hager et al, JCP (2016)].
- Sometimes exact energy conservation could be important, e.g. mitigate numerical heating in edge pedestal in long simulations: Example case, numerical electron heating up to 6 % of experimental input power.

Radially integrated numerical heating power in the full-current ITER simulation



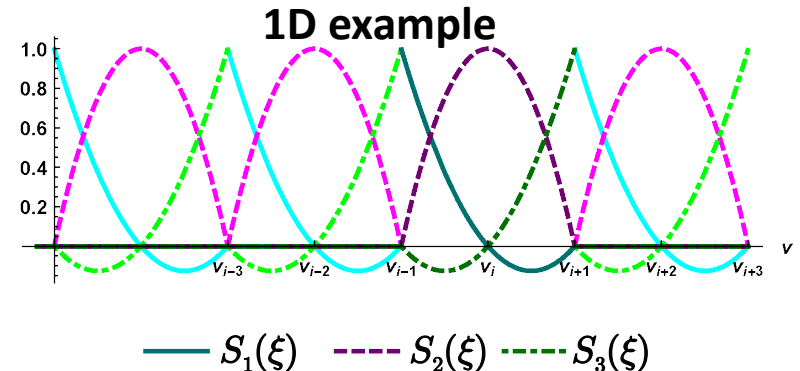
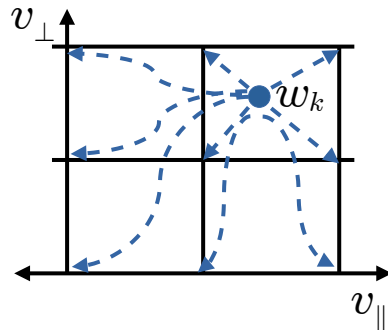
- XGC implements a bilinear mapping in velocity space [Yoon, Chang, PPCF (2014)].
 - Marker particle weights are transferred to the grid using linear shape functions. \Rightarrow Fraction \propto inverse linear distance between particle and surrounding 4 grid points.
 - Conserves particle number and momentum but not energy at the same level.

1st order



- We need quadratic mapping (or higher order) to also conserve energy.

2nd order



- For particles \rightarrow grid mapping:

Quadratic Lagrange P_2 shape functions

$$S_1(\xi) = 1 - 3\xi + 2\xi^2$$

$$S_2(\xi) = 4\xi - 4\xi^2$$

$$S_3(\xi) = -\xi + 2\xi^2$$

- 1D velocity space: One element consists of 3 grid points.

Let $\xi(v) = (v - v_{i-1}) / (v_{i+1} - v_{i-1}) \in [0, 1]$.

$S_1(\xi)$, $S_2(\xi)$, $S_3(\xi)$ should fulfill:

Particle conservation $S_1(\xi) + S_2(\xi) + S_3(\xi) = 1$

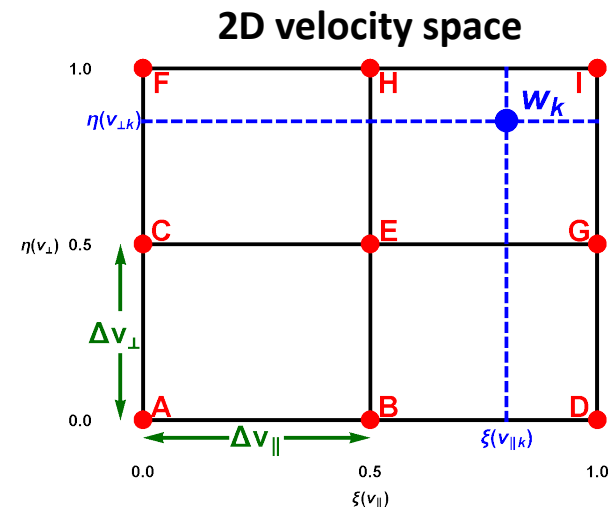
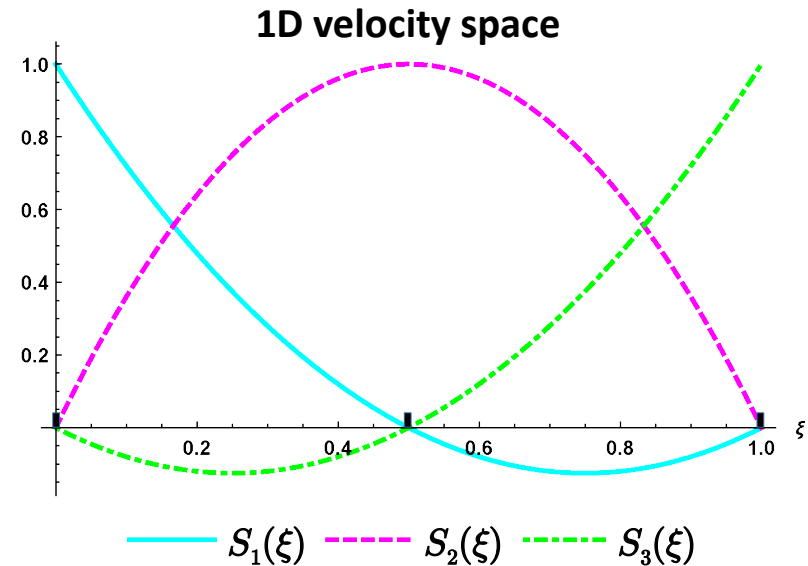
Momentum conservation $0 \cdot S_1(\xi) + 0.5 \cdot S_2(\xi) + 1.0 \cdot S_3(\xi) = \xi$

Energy conservation $0^2 \cdot S_1(\xi) + 0.5^2 \cdot S_2(\xi) + 1.0^2 \cdot S_3(\xi) = \xi^2$

- More difficult to do the inverse grid \rightarrow particles mapping.

We want to preserve moments \leq order of elements.

Solution: mapping with pseudo-inverse.



[Hirvijoki et al, arXiv:1802.05263 (2018)]

- Particle distribution of $N_{\text{particles}}$ particles
Finite element space \mathcal{F} and a function $u \in \mathcal{F}$
- $u = f_{\text{particle}}(\mathbf{v})$ generally not possible
For basis functions $\phi_j \in \mathcal{F}$ we write
Solution c_j should fulfill weak equivalence
- Mass matrix $M_{ij} = \int d\mathbf{v} \phi_i \phi_j$ [$N_{\text{vertices}} \times N_{\text{vertices}}$]
Coefficients vector \mathbf{c}

Weak equivalence

- Particles \rightarrow grid mapping
On grid do operations (e.g. collisions)
 V rectangular so inverse not well defined \Rightarrow
E.g. if $N_{\text{particles}} > N_{\text{vertices}}$ use right inverse

- Use XGC resampling to ensure pseudo-inverse calculation (good particle coverage in phase-space).

$$f_{\text{particle}}(\mathbf{v}) = \sum_k w_k \delta(\mathbf{v} - \mathbf{v}_k)$$

$$u = \sum_j c_j \phi_j$$

$$\int d\mathbf{v} \phi_i f_{\text{particle}} = \int d\mathbf{v} \phi_i u$$

$$\text{"Positions" matrix } V_{jk} = \phi_j(\mathbf{v}_k) \text{ } [N_{\text{vertices}} \times N_{\text{particles}}]$$

$$\text{Weights vector } \mathbf{w}$$

$$M\mathbf{c} = V\mathbf{w} \quad \Rightarrow \quad \mathbf{c} = M^{-1}V\mathbf{w}$$

$$f_{\text{grid}} = V\mathbf{w} \quad (= M\mathbf{c})$$

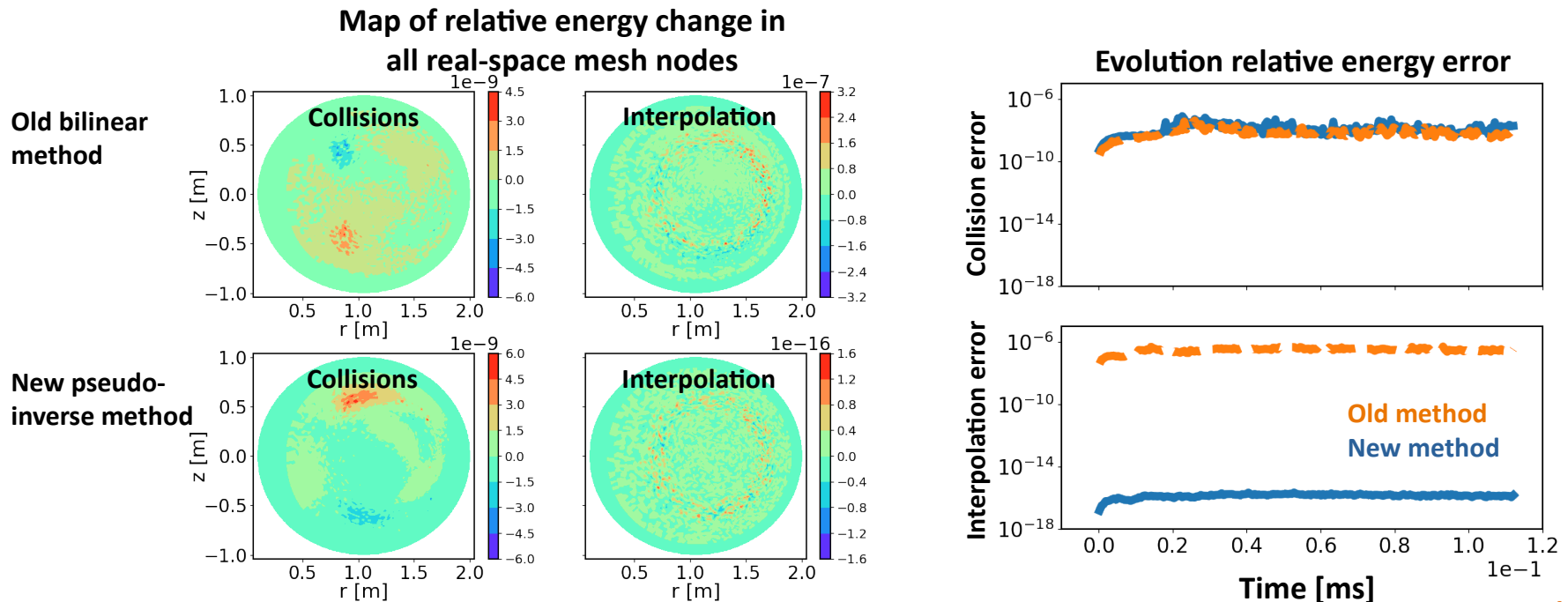
$$\mathbf{c} \quad \rightarrow \quad \mathbf{c}_{\text{new}}$$

grid \rightarrow particles mapping with pseudo-inverse V^+

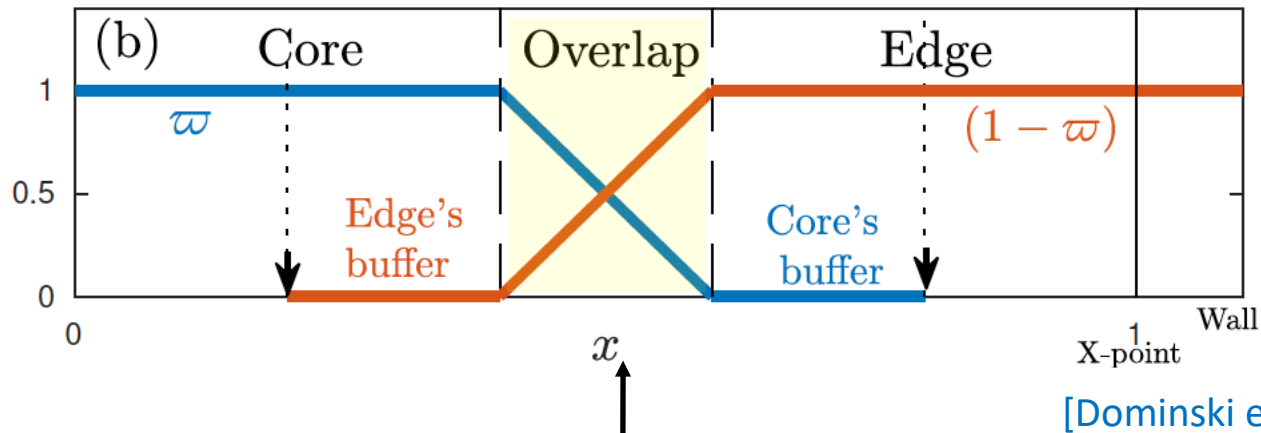
$$\mathbf{w}_{\text{new}} = V^T(VV^T)^{-1}(M\mathbf{c}_{\text{new}})$$

- Fokker-Planck-Landau collision operator in XGC conserves particle density, total parallel momentum and total kinetic energy.
- We want interpolation error \ll collision error $<$ tolerance.
- Collision and pseudo-inverse interpolation errors can be made arbitrarily small by stricter tolerances in iterative solvers. Bilinear interpolation errors set by the algorithm, e.g. does not conserve energy.

Circular tokamak example



Necessity for more accurate resampling and v-space interpolation in WDMApp f-coupling



[Dominski et al, PoP (2021)]

Need to construct a common $f(x, v)$ in the overlap region

- GENE - XGC: XGC's $f(x, v)$ needs to be smooth. Otherwise, Courant-Friedrichs-Lewy condition can be an issue in GENE. Also, GENE's $f(x, v)$ may need to be converted to XGC's particles.
- GEM - XGC: delta- f GEM uses much less particles/vertex than XGC. Particle number needs to be up/down sampled with accurate velocity-space interpolation for the good common $f(x, v)$.
- A common Fokker-Planck-Landau operation, without an artificial heating, is needed in the coupler for the overlap region. Different $C(f)$ operator in each code can yield different $f(x, v)$.

- Motivation
- Brief introduction to the XGC code
- Particle resampling in a PIC code
- Updated velocity-space interpolation using a pseudo-inverse
- Summary

- Particle resampling and pseudo-inverse velocity mapping are two techniques that will help enable long-time simulations of High-confinement mode plasmas with steep edge pedestals.
- Particle resampling is a powerful tool in XGC with several applications.
- New pseudo-inverse velocity mapping implemented in XGC works as intended: conserves particle number, momentum and energy exactly.
- Both collision error and interpolation error can now be made arbitrarily small down to machine precision by choosing stricter tolerances in the iterative solvers.

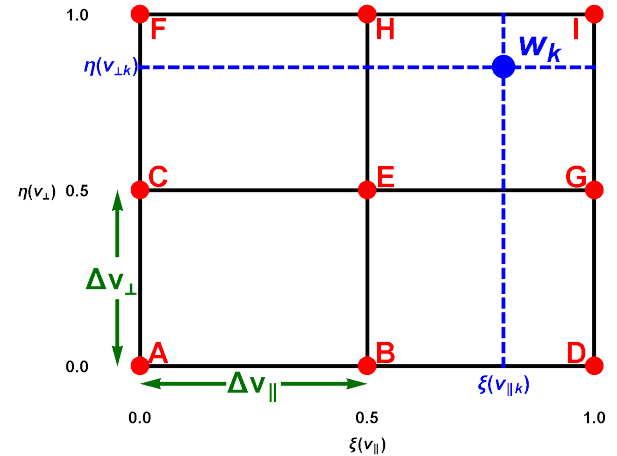
- 2D uniform rectangular grid in v_{\parallel}, v_{\perp} .
 $\xi(v_{\parallel}) = (v_{\parallel} - v_{\parallel, i-1}) / (2\Delta v_{\parallel})$; $\eta(v_{\perp}) = (v_{\perp} - v_{\perp, j-1}) / (2\Delta v_{\perp})$
- Map marker particles (with weights w_k) to the 9 vertices using

$$\mathcal{P}_{k \rightarrow A} = S_1(\xi(v_{\parallel k})) S_1(\eta(v_{\perp k}))$$

$$\mathcal{P}_{k \rightarrow B} = S_2(\xi(v_{\parallel k})) S_1(\eta(v_{\perp k}))$$

\vdots

$$\mathcal{P}_{k \rightarrow I} = S_3(\xi(v_{\parallel k})) S_3(\eta(v_{\perp k}))$$



- Distribution on the mesh $f_{\text{mesh}, \beta} = V_{\beta}^{-1} \sum_k w_k \mathcal{P}_{k \rightarrow \beta}$; V_{β} - velocity space volume of vertex β .
- After source operations, inverse mapping back to particle α
 $\mathcal{P}_{\beta \rightarrow \alpha}^{(-1)} = \mathcal{P}_{\alpha \rightarrow \beta} / (\sum_k \mathcal{P}_{k \rightarrow \beta})$; New weight $w_{\alpha}^{\text{new}} = \sum_{\beta} f_{\text{mesh}, \beta}^{\text{new}} V_{\beta} \mathcal{P}_{\beta \rightarrow \alpha}^{(-1)}$.
- Shape functions $S_1(\xi)$, $S_3(\xi)$ are negative in parts of the domain \Rightarrow
Mapping coefficients $\mathcal{P}_{k \rightarrow \beta}$ can be negative \Rightarrow
Problem! Risk of division by 0 in inverse grid \rightarrow particles mapping.

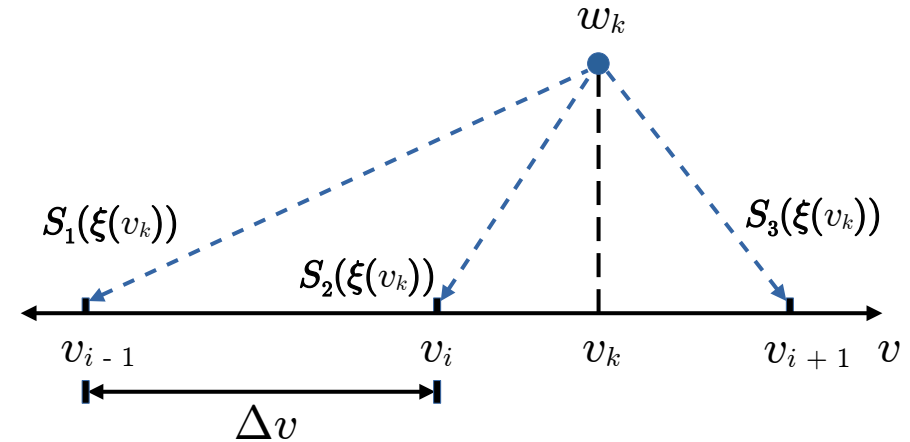
Quadratic Lagrange P_2 shape functions

$$S_1(\xi) = 1 - 3\xi + 2\xi^2$$

$$S_2(\xi) = 4\xi - 4\xi^2$$

$$S_3(\xi) = -\xi + 2\xi^2$$

$$\xi(v) = (v - v_{i-1}) / (2\Delta v) \in [0, 1]$$



Particle conservation

$$n_{\text{particle}, k} = w_k$$

$$n_{\text{grid}, i-1} + n_{\text{grid}, i} + n_{\text{grid}, i+1} = w_k S_1(\xi(v_k)) + w_k S_2(\xi(v_k)) + w_k S_3(\xi(v_k)) = w_k$$

Momentum conservation

$$P_{\text{particle}, k} = w_k v_k$$

$$P_{\text{grid}, i-1} + P_{\text{grid}, i} + P_{\text{grid}, i+1} = w_k S_1(\xi(v_k)) v_{i-1} + w_k S_2(\xi(v_k)) v_i + w_k S_3(\xi(v_k)) v_{i+1} = w_k v_k$$

Energy conservation

$$E_{\text{particle}, k} = w_k v_k^2$$

$$E_{\text{grid}, i-1} + E_{\text{grid}, i} + E_{\text{grid}, i+1} = w_k S_1(\xi(v_k)) v_{i-1}^2 + w_k S_2(\xi(v_k)) v_i^2 + w_k S_3(\xi(v_k)) v_{i+1}^2 = w_k v_k^2$$

If $N_{\text{particles}} < N_{\text{vertices}}$ (unusual case) use left pseudo-inverse:

- Mass matrix $M_{ij} = \int d\mathbf{v} \phi_i \phi_j$

Particle positions matrix $V_{jk} = \phi_j(\mathbf{v}_k) = \begin{pmatrix} * & * & * \\ * & * & * \\ * & * & * \\ * & * & * \end{pmatrix}$

Weak equivalence

- Particles \rightarrow grid mapping
On grid do operations (e.g. collisions)
Left pseudo-inverse ($V^+ V = I$)

New weights (preserves moments \leq order)

If identity operation on grid:

invertible square matrix size $N_{\text{vertices}} \times N_{\text{vertices}}$

tall skinny matrix size $N_{\text{vertices}} \times N_{\text{particles}}$

$$M\mathbf{c} = V\mathbf{w}$$

$$f_{\text{grid}} = V\mathbf{w} \quad (= M\mathbf{c})$$

$$\mathbf{c} \rightarrow \mathbf{c}_{\text{new}}$$

$$V^+ = (V^T V)^{-1} V^T$$

$$\mathbf{w}_{\text{new}} = (V^T V)^{-1} V^T (M\mathbf{c}_{\text{new}})$$

$$\mathbf{w}_{\text{new}} = (V^T V)^{-1} V^T (V\mathbf{w}) = \mathbf{w}$$

If $N_{\text{particles}} > N_{\text{vertices}}$ (typical case) use right (Moore-Penrose) pseudo-inverse:

- Mass matrix $M_{ij} = \int d\mathbf{v} \phi_i \phi_j$

invertible square matrix size $N_{\text{vertices}} \times N_{\text{vertices}}$

Particle positions matrix $V_{jk} = \phi_j(\mathbf{v}_k) = \begin{pmatrix} * & * & * & * & * \\ * & * & * & * & * \\ * & * & * & * & * \end{pmatrix}$ short fat matrix size $N_{\text{vertices}} \times N_{\text{particles}}$

Weak equivalence

$$M\mathbf{c} = V\mathbf{w}$$

- Particles \rightarrow grid mapping

On grid do operations (e.g. collisions)

Right pseudo-inverse ($VV^+ = I$)

$$f_{\text{grid}} = V\mathbf{w} \quad (= M\mathbf{c})$$

$$\mathbf{c} \rightarrow \mathbf{c}_{\text{new}}$$

$$V^+ = V^T(VV^T)^{-1}$$

New weights (preserves moments \leq order)

$$\mathbf{w}_{\text{new}} = V^T(VV^T)^{-1}(M\mathbf{c}_{\text{new}})$$

If identity operation on grid:

$$\mathbf{w}_{\text{new}} = V^T(VV^T)^{-1}(V\mathbf{w}) \neq \mathbf{w}$$

Idempotence (map twice = map once):

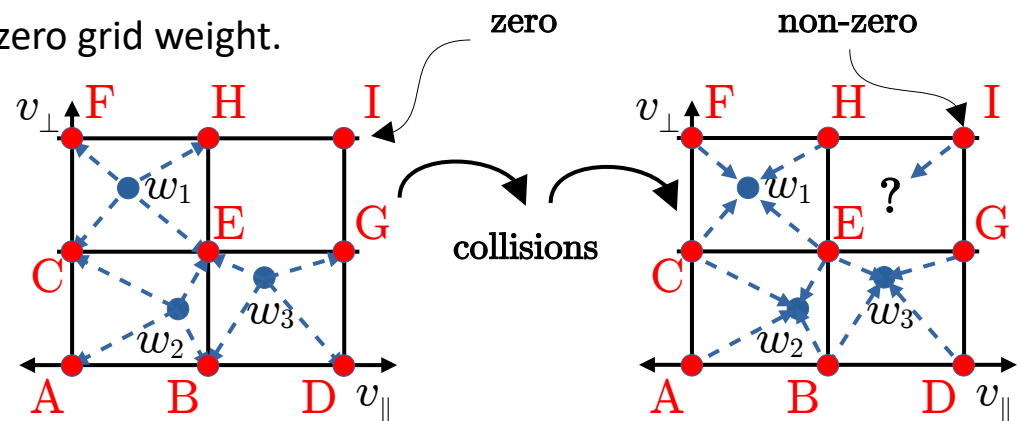
$$V^T(VV^T)^{-1}VV^T(VV^T)^{-1}V\mathbf{w} = V^T(VV^T)^{-1}V\mathbf{w}$$

We need the resampling in XGC for the velocity mapping

To ensure that pseudo-inverse calculation always work all basis functions (cells) must have particles.

Example

- For simplicity assume 1st order bilinear mapping, 3 particles in 4 cells.
- After particles → grid mapping vertex **I** has 0 grid weight.
- After collision operation vertex **I** has non-zero grid weight.
- Vertex **I** has no particle to write back grid weight to in grid → particles mapping.

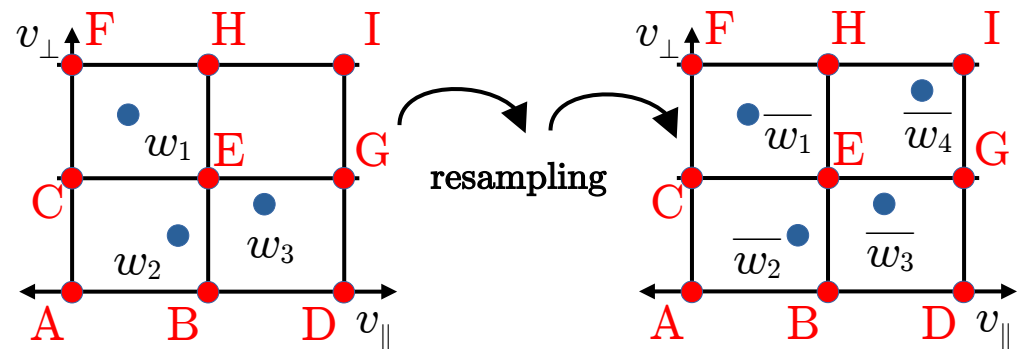


Solution

- Use resampling in XGC to fill all cells before particles → grid mapping.

[Faghihi et al, JCP (2020)]

[Dominski et al, PoP (2020)]



- Particle positions and weights arrays are constructed in the Fortran part of XGC.
- Interface to C++ where PETSc interpolation routines are used.
- Pseudo-inverse calculated with iterative Krylov solver in PETSc.
- At this point the new interpolation is slow. Bottlenecks are to construct the M_{ij} and V_{jk} matrices.
- There is room for substantial speedup:
 - Besides MPI implement OpenMP parallelization (no communication between ranks/threads).
 - The code can be rewritten to be more efficient.
 - Rewrite all XGC source routines into C++ with Kokkos?
 - Use a different collision operator with adaptive mesh refinement (AMR) [\[M. F. Adams\]](#).

AMR example

